

## DAY 4

# Discovery of genomic structural variations using NGS data

26.11.2015

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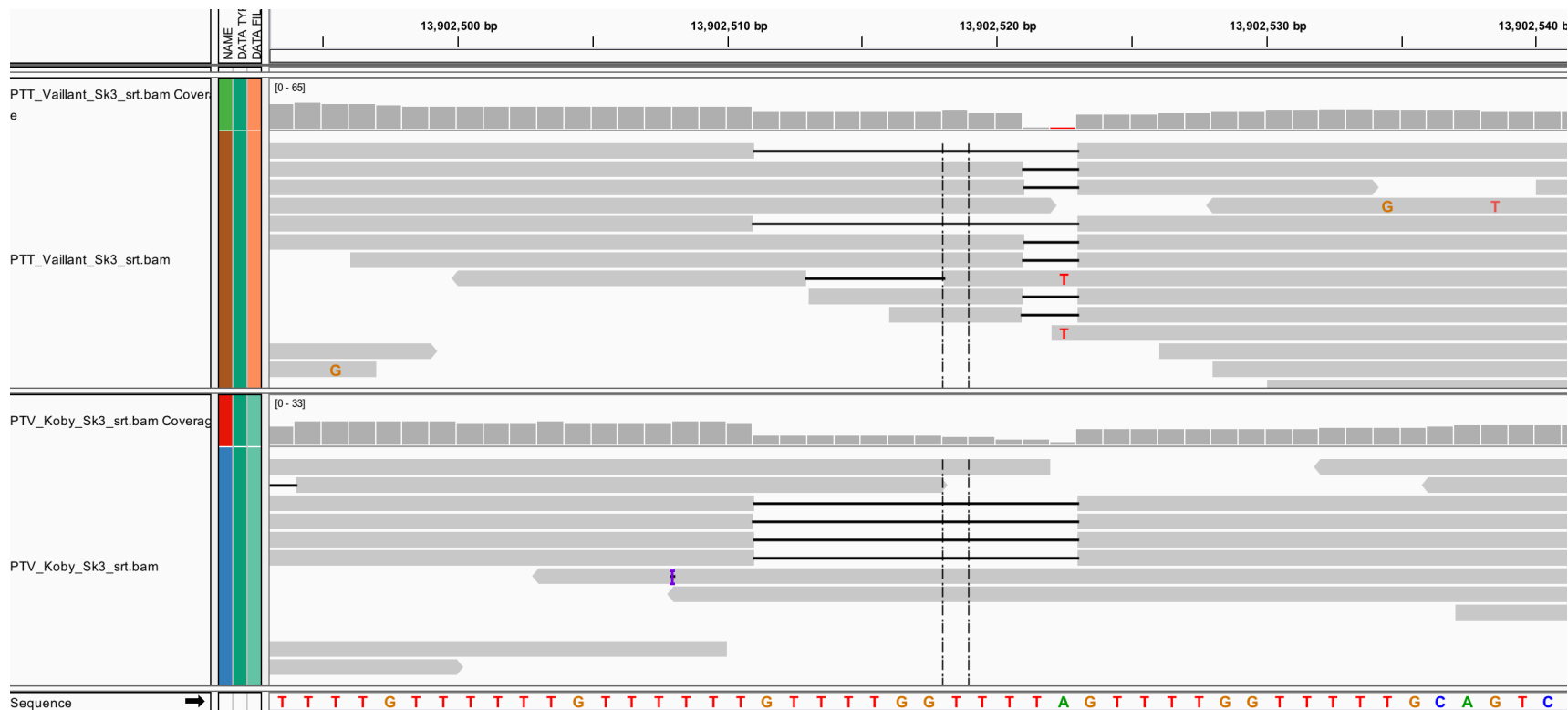


# Overview

- Small indels
- Genomic rearrangements/structural variations (SV)
- Discovery of SVs
  - Other methods used so far
  - Using NGS data:
    - Whole genomes
    - Capture data (e.g. exomes, custom capture)
    - other
- Practical using software pindel

# Small indels

- 1 bp to a few tens of bp (in this context)
- Calling reliability depends on sequencing technology and type of data
- Most variant callers include small indel calling
- More care is needed (e.g. overlap between different callers)
- Validation!!



# variant calling – VCF file

```
##fileformat=VCFv4.1
##samtoolsVersion=0.1.18 (r982:295)
##INFO=<ID=DP,Number=1,Type=Integer,Description="Raw read depth">
##INFO=<ID=DP4,Number=4,Type=Integer,Description="# high-quality ref-forward bases, ref-reverse, alt-forward and alt-reverse bases">
##INFO=<ID=MQ,Number=1,Type=Integer,Description="Root-mean-square mapping quality of covering reads">
##INFO=<ID=FQ,Number=1,Type=Float,Description="Phred probability of all samples being the same">
##INFO=<ID=AF1,Number=1,Type=Float,Description="Max-likelihood estimate of the first ALT allele frequency (assuming HWE)">
##INFO=<ID=AC1,Number=1,Type=Float,Description="Max-likelihood estimate of the first ALT allele count (no HWE assumption)">
##INFO=<ID=G3,Number=3,Type=Float,Description="ML estimate of genotype frequencies">
##INFO=<ID=HWE,Number=1,Type=Float,Description="Chi^2 based HWE test P-value based on G3">
##INFO=<ID=CLR,Number=1,Type=Integer,Description="Log ratio of genotype likelihoods with and without the constraint">
##INFO=<ID=UGT,Number=1,Type=String,Description="The most probable unconstrained genotype configuration in the trio">
##INFO=<ID=CGT,Number=1,Type=String,Description="The most probable constrained genotype configuration in the trio">
##INFO=<ID=PV4,Number=4,Type=Float,Description="P-values for strand bias, baseQ bias, mapQ bias and tail distance bias">
##INFO=<ID=INDEL,Number=0,Type=Flag,Description="Indicates that the variant is an INDEL.">
##INFO=<ID=PC2,Number=2,Type=Integer,Description="Phred probability of the nonRef allele frequency in group1 samples being larger (.smaller) than i">
##INFO=<ID=PCHI2,Number=1,Type=Float,Description="Posterior weighted chi^2 P-value for testing the association between group1 and group2 samples.">
##INFO=<ID=QCHI2,Number=1,Type=Integer,Description="Phred scaled PCHI2.">
##INFO=<ID=PR,Number=1,Type=Integer,Description="# permutations yielding a smaller PCHI2.">
##INFO=<ID=VDB,Number=1,Type=Float,Description="Variant Distance Bias">
##FORMAT=<ID=GT,Number=1,Type=String,Description="Genotype">
##FORMAT=<ID=GQ,Number=1,Type=Integer,Description="Genotype Quality">
##FORMAT=<ID=GL,Number=3,Type=Float,Description="Likelihoods for RR,RA,AA genotypes (R=ref,A=alt)">
##FORMAT=<ID=DP,Number=1,Type=Integer,Description="# high-quality bases">
##FORMAT=<ID=SP,Number=1,Type=Integer,Description="Phred-scaled strand bias P-value">
##FORMAT=<ID=PL,Number=G,Type=Integer,Description="List of Phred-scaled genotype likelihoods">
##FILTER=<ID=StrandBias,Description="Min P-value for strand bias (INFO/PV4) [0.0001]">
##FILTER=<ID=EndDistBias,Description="Min P-value for end distance bias (INFO/PV4) [0.0001]">
##FILTER=<ID=MaxDP,Description="Maximum read depth (INFO/DP or INFO/DP4) [10000000]">
##FILTER=<ID=BaseQualBias,Description="Min P-value for baseQ bias (INFO/PV4) [0]">
##FILTER=<ID=MinMQ,Description="Minimum RMS mapping quality for SNPs (INFO/MQ) [10]">
##FILTER=<ID=Qual,Description="Minimum value of the QUAL field [10]">
##FILTER=<ID=MinAB,Description="Minimum number of alternate bases (INFO/DP4) [2]">
##FILTER=<ID=VDB,Description="Minimum Variant Distance Bias (INFO/VDB) [0.015]">
##FILTER=<ID=GapWin,Description="Window size for filtering adjacent gaps [3]">
##FILTER=<ID=MapQualBias,Description="Min P-value for mapQ bias (INFO/PV4) [0]">
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##FILTER=<ID=MinDP,Description="Minimum read depth (INFO/DP or INFO/DP4) [2]">
##FILTER=<ID=RefN,Description="Reference base is N []">
##FILTER=<ID=HWE,Description="Minimum P-value for HWE (plus F<0) (INFO/HWE and INFO/G3) [0.0001]">
##source_20130519.1=/usr/bin/vcf-annotate -f +
#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT 60A-Sc-DBVPG6044
I 2709 . G T 74.1 PASS DP=8;VDB=0.0321;AF1=1;AC1=2;DP4=0,0,6,1;MQ=60;FQ=-48 GT:PL:GQ 1/1:107,21,0:39
I 2825 . G C 73.3 PASS DP=5;VDB=0.0302;AF1=1;AC1=2;DP4=0,0,1,4;MQ=60;FQ=-42 GT:PL:GQ 1/1:106,15,0:27
I 2875 . TAA TAAA 96.5 PASS INDEL;DP=11;VDB=0.0321;AF1=0.5;AC1=1;DP4=3,1,1,5;MQ=60;FQ=44.5;PV4=0.19,1,1,0.041 GT: 0/1
I 2891 . G T 156 PASS DP=12;VDB=0.0280;AF1=1;AC1=2;DP4=0,0,5,6;MQ=60;FQ=-60 GT:PL:GQ 1/1:189,33,0:63
I 2914 . G A 96 PASS DP=12;VDB=0.0280;AF1=0.5;AC1=1;DP4=2,2,2,5;MQ=60;FQ=57;PV4=0.58,0.0066,1,0.43 GT:PL:GQ 0/1
I 3022 . C G 23 VDB DP=5;VDB=0.0135;AF1=0.5;AC1=1;DP4=2,1,1,1;MQ=60;FQ=26;PV4=1,1,1,1 GT:PL:GQ 0/1
I 3106 . T A 15.1 VDB DP=7;VDB=0.0135;AF1=0.5;AC1=1;DP4=3,2,1,1;MQ=60;FQ=18.1;PV4=1,0.29,1,1 GT:PL:GQ 0/1
I 3197 . G A 26 VDB DP=7;VDB=0.0112;AF1=0.5;AC1=1;DP4=3,0,1,3;MQ=60;FQ=28.2;PV4=0.14,6.3e-06,1,1 GT:PL:GQ 0/1
I 3226 . G A 9.52 Qual;VDB DP=8;VDB=0.0112;AF1=0.5;AC1=1;DP4=2,2,1,2;MQ=60;FQ=12.3;PV4=1,9.5e-08,1,1 GT: 0/1
I 3688 . G A 21 PASS DP=9;VDB=0.0240;AF1=0.5;AC1=1;DP4=2,2,1,3;MQ=60;FQ=24;PV4=1,1.7e-08,1,1 GT:PL:GQ 0/1
```

- Which one is the indel?
- What type of change is it?

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OPEN

doi:10.1038/nature15393

# A global reference for human genetic variation

The 1000 Genomes Project Consortium\*

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OPEN

doi:10.1038/nature15394

# An integrated map of structural variation in 2,504 human genomes

A list of authors and their affiliations appears at the end of the paper.

Nature 1 October 2015

	<b>Autosomes</b>	<b>Exome target regions**</b>	<b>chrX***</b>	<b>chrY***</b>	<b>Totals</b>
<b>Samples</b>	2,504	2,504	2,504	1,233	-
<b>Total Raw Bases (Gb)</b>	85,426	18,273	3,213	291	-
<b>Mean Mapped Depth (X)*</b>	8.45	75.25	6.20	2.60	-
<b>Total Variant Sites</b>	84,801,880	1,416,049	3,468,093	62,042	88,332,015
<b>Biallelic SNPs</b>	81,102,777	1,383,927	3,223,927	60,505	84,387,209
<b>Indels</b>	3,196,364	19,832	212,196	1,427	3,409,987
<b>Mean Indel Length (bp)</b>	2.94	3.46	2.64	2.00	-
<b>Multiallelic sites</b>	444,026	6,153	30,996	-	475,022
<b>Multiallelic SNPs</b>	274,425	4,706	15,055	-	289,480
<b>Multiallelic Indels</b>	169,601	1,447	15,941	-	185,542
<b>Structural Variants</b>	58,713	6,137	974	110	59,797

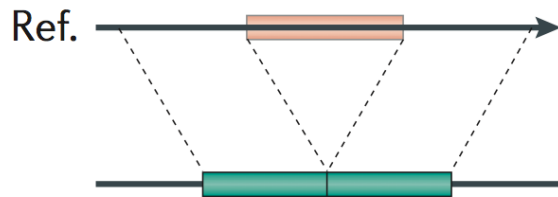
From 1000G phase 3 (Nature 2015 Oct 1)

# Structural variations (SVs)

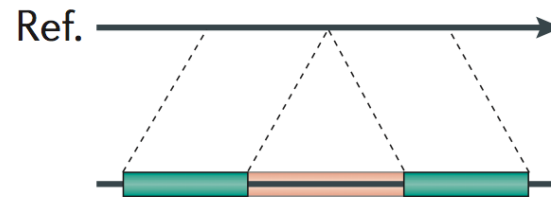
- Medium events (>10 bp up to 1 kb)
- Large events (typically >1kb to many Mb)
- Affect more bases than SNPs/short indels
  - 4.1 – 5M variant sites compared to the reference:
    - >99.9% - SNPs and short indels
    - 2,100 – 2,500 structural variants – affect ~20Mb!!

# What are they?

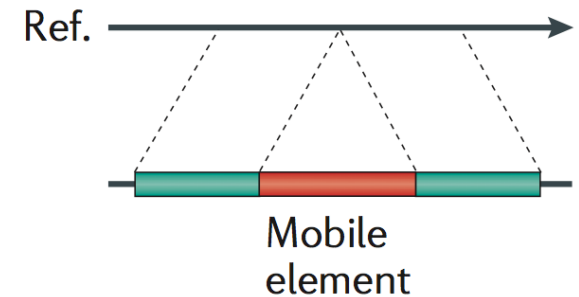
## Deletion



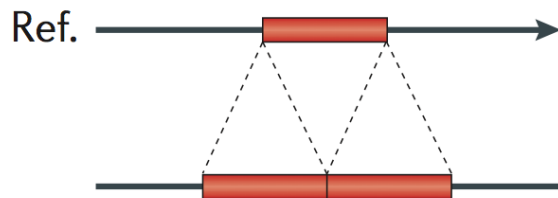
## Novel sequence insertion



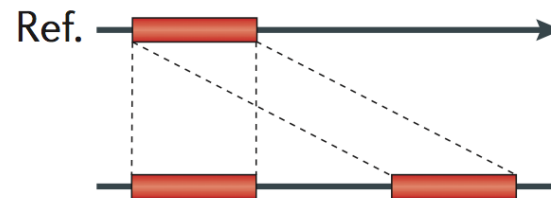
## Mobile-element insertion



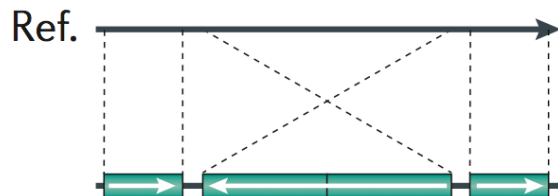
## Tandem duplication



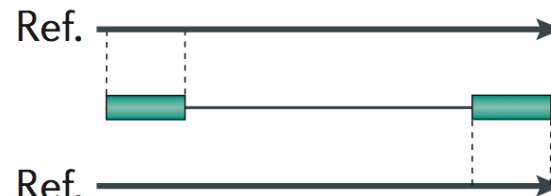
## Interspersed duplication



## Inversion



## Translocation





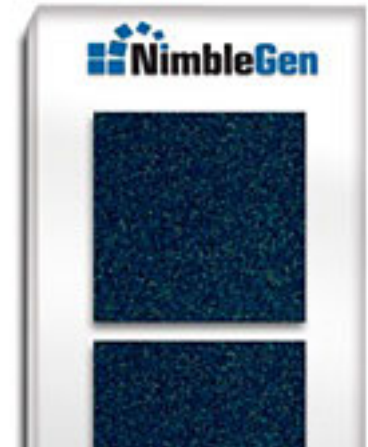
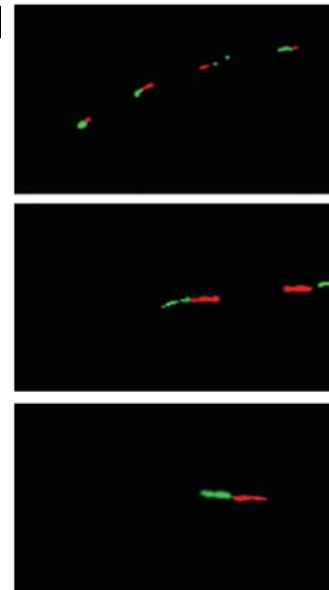
# Why study?

- Can have severe functional consequences as large genomic regions are affected
- Gene dosage/ gene disruption/ position/ unmasking a recessive allele etc
  - Charcot–Marie–Tooth disease type 1A – 1.4Mb duplication
  - hereditary neuropathy – 1.4Mb deletion
  - Smith–Magenis microdeletion syndrome – 3.7Mb deletion
  - Sex-reversal – *SRY* chrY to chrX translocation
  - hemophilia A – inversion in factor VIII gene
  - Fragile X syndrome and Huntington's disease – trinucleotide repeat expansion
  - salivary amylase (*AMY1*) – digestion of starch, 1-10 copies

} 17p11.2p12

# Methods used so far

- Fluorescent in situ hybridisation (FISH)
  - Fluorescent probes (~100kb)
  - Detect presence or absence of a specific region
- Array Comparative Genomic Hybridisation (ACGH)
  - Test vs reference sample
  - Millions of probes, defining the resolution (ultra-high resolution 24-42M probes)
- SNP arrays
  - Single sample per microarray
  - Millions of probes, defining the resolution
  - Offer SNP genotype information

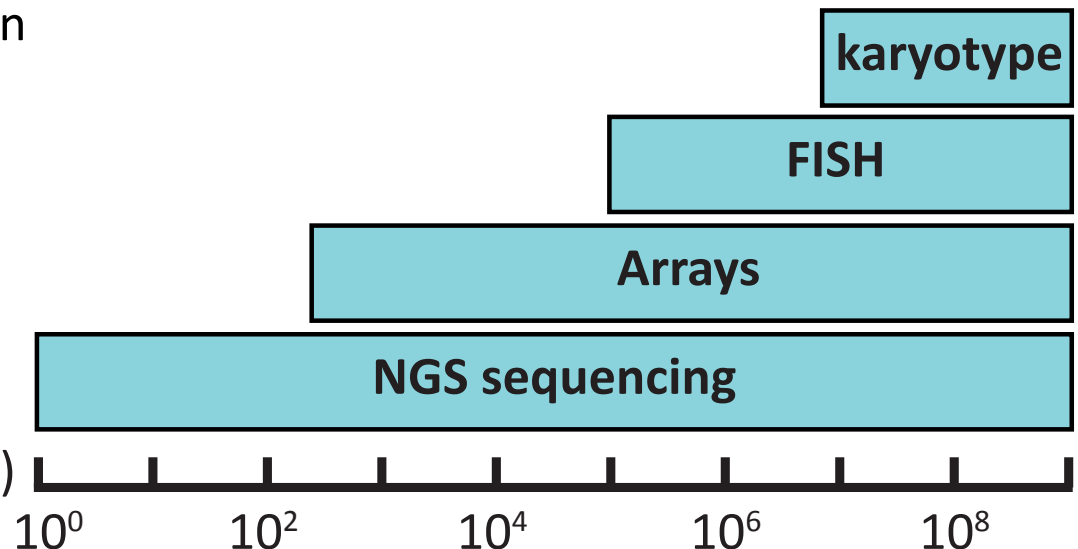


- Limitations of arrays

- Dependent on the reference sequence used to design the probes
- Lower resolution
- No information about location
- Cannot detect balanced events (e.g. inversions)
- Performance in repeat-rich and duplicated regions is not great
- No breakpoint information

- Advantages of arrays

- Cheaper
- Faster
- High throughput



- Difficulties created by NGS data:
  - Quality of indel calls depend on the technology used
  - Relatively short read-lengths
  - Insert size
  - Sequence coverage
- Advantages of NGS data:
  - Possibility to detect different variants in a single experiment
  - Largely unbiased
  - Complete spectrum of genetic variation
  - Breakpoint information

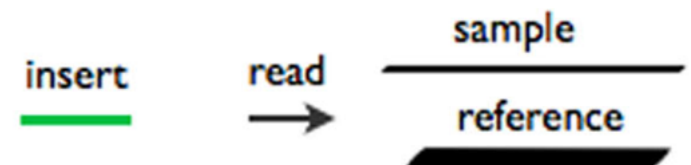
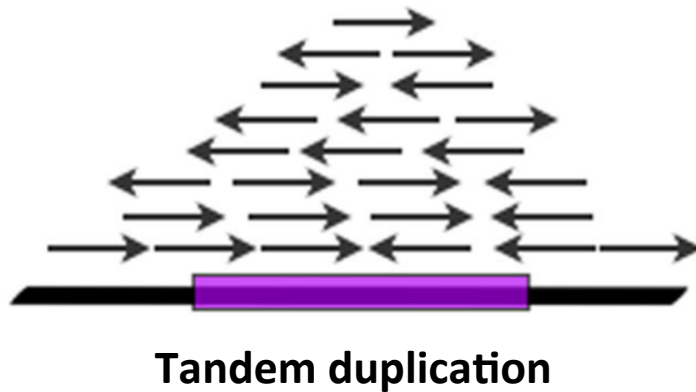
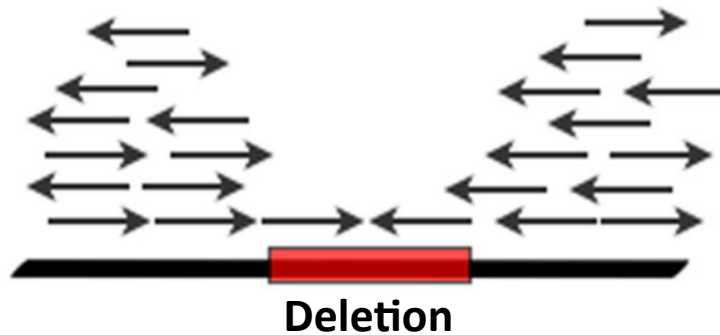
# NGS data: methods

- Read depth
- Read-pair
- Split-read
- *De novo* assembly
- Combination of the above

# WGS data

- Read depth methods:
  - Assess read depth (higher DP – higher copy number)
  - Assume random (typically Poisson or modified Poisson) distribution of read depth
  - Study design for normalisation: single-sample; paired case/control; large population sample
  - Divergence from the distribution to identify:
    - Deletions – significantly reduced DP
    - Duplications – significantly higher DP
  - Absolute copy-number prediction
  - Usually poor break-point resolution
  - E.g. ReadDepth, mrCaNaVar, RDXplorer, CNV-seq, cn.MOPS and CNVnator

# WGS data: read depth methods



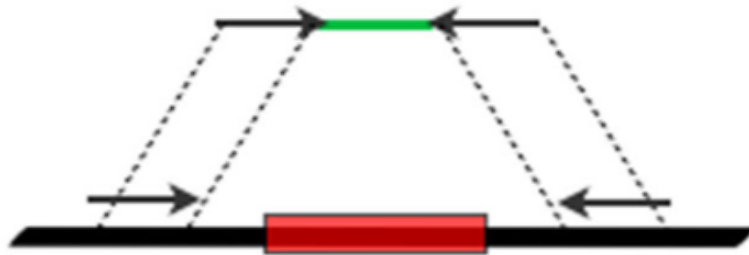
Modified from Tattini et al. 2015

# WGS data

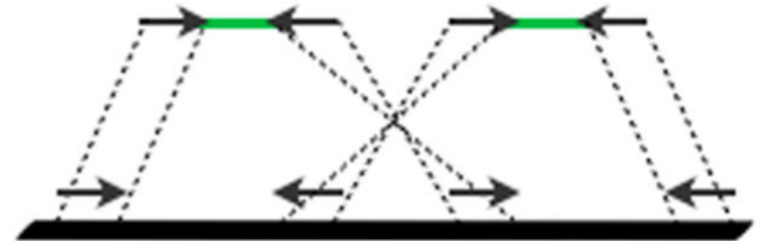
- Read-pair methods:
  - Assess the span and orientation of paired-end reads
  - Require tight insert size distribution
  - Find discordant pairs where mapping span and/or orientation is discordant with reference genome
    - Deletions – mapping too far
    - Insertions – mapping too close
    - Inversions and some tandem duplications – orientation inconsistency
  - Can identify a wide range of rearrangements
  - Do not perform well in repetitive regions
  - Breakpoint prediction depends on very tight fragment size distribution
  - E.g. PEMer, VariationHunter, BreakDancer, MoDIL, MoGUL, HYDRA, Corona and SPANNER



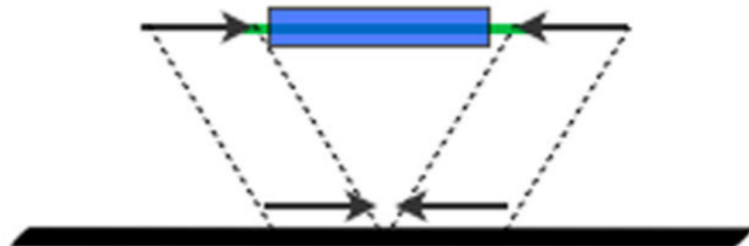
# WGS data: read-pair methods



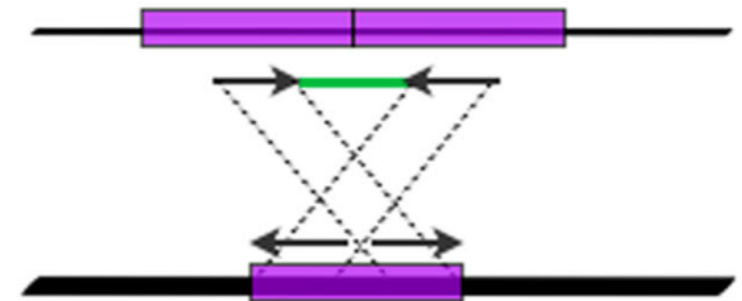
Deletion



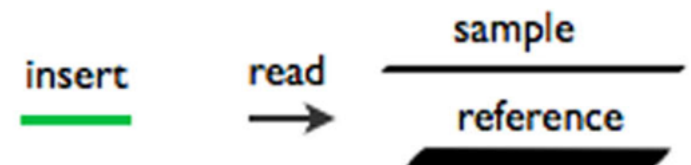
Inversion



Insertion



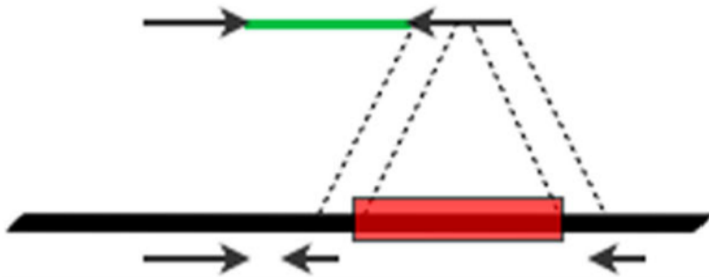
Tandem duplication



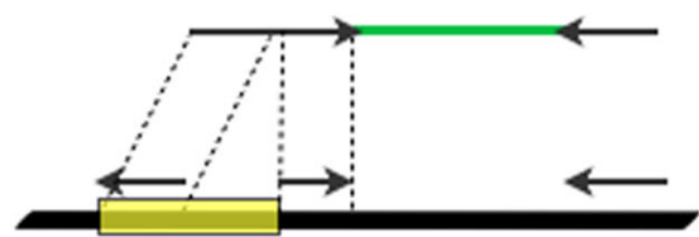
# WGS data

- Split-read methods:
  - Identify the breakpoint of a SV based on a “split” in the read signature
  - One read maps uniquely, while the other fails to map or maps partly
  - Detection of a number of rearrangements
    - May be able to detect mobile-element insertions
  - Exact breakpoint resolution
  - Relies heavily on read length, but also insert size
  - Reliable only in unique parts of the genome
  - E.g. Pindel, AGE, Splitread, SLOPE, SRiC

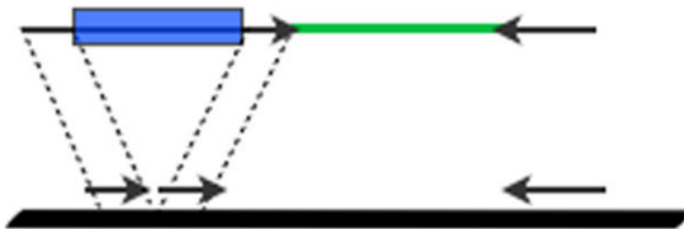
# WGS data: split-read methods



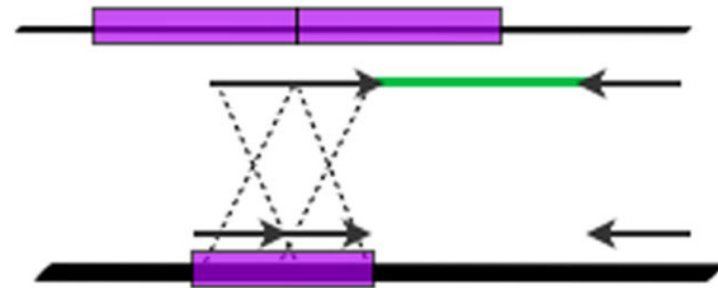
Deletion



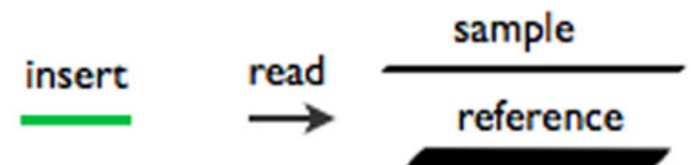
Inversion



Insertion

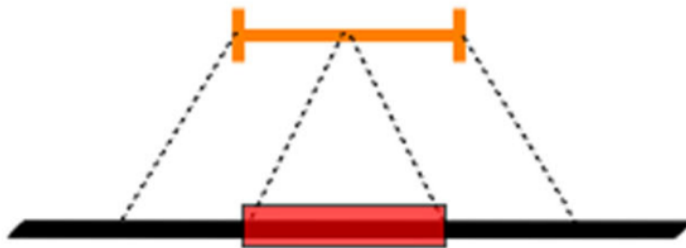


Tandem duplication

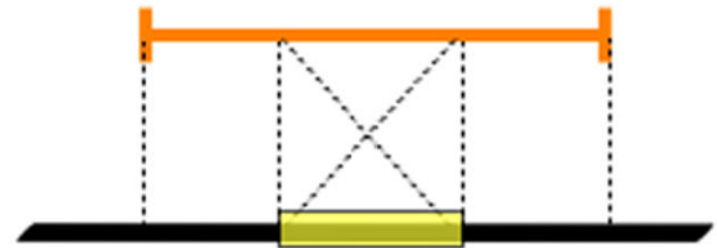


# WG-sequence data

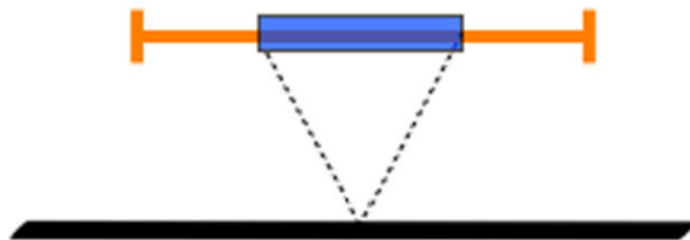
- Sequence assembly:
  - *De novo* assembly of the sequence followed by comparison to a reference genome
    - should allow accurate typing of copy, content and structure
  - Often use combination of *de novo* and local assembly algorithms
  - Most promising of all the methods
  - Biased against repeats and duplications
  - E.g.
    - *de novo* assembly: EULER-USR, ABySS, SOAPdenovo and ALLPATHS-LG
    - Mixed approaches: Cortex, NovelSeq, TIGRA



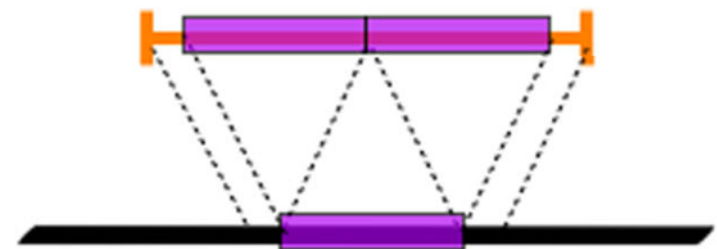
**Deletion**



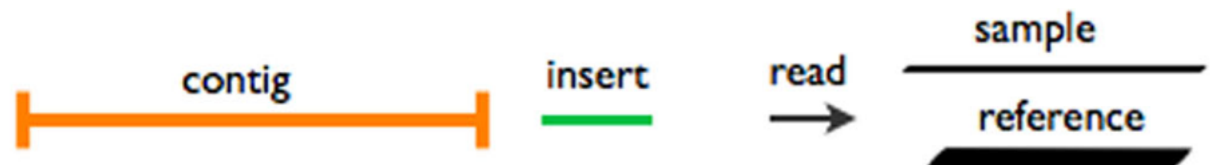
**Inversion**



**Insertion**

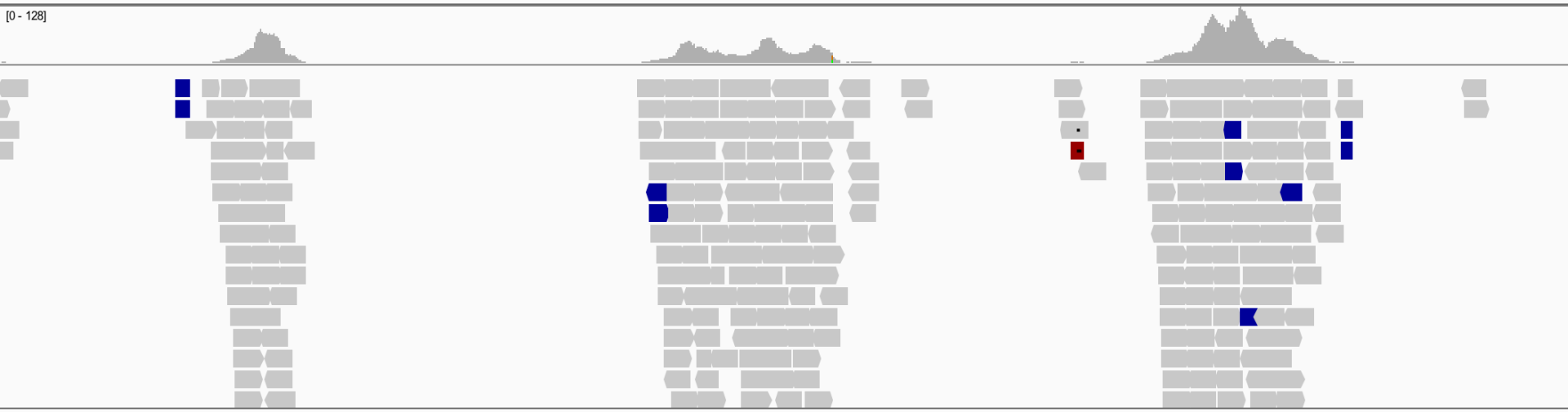
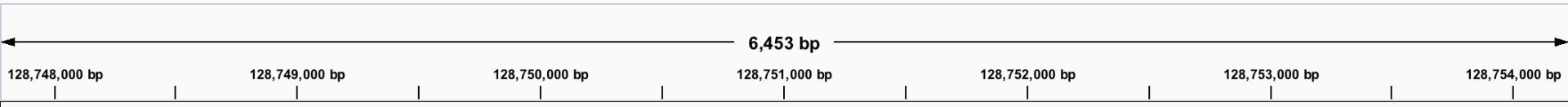
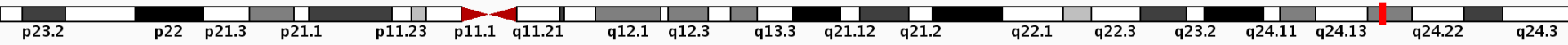


**Tandem duplication**



# Exome or custom capture data

- Characteristics of the data
  - Uneven coverage, but usually much higher depth
  - Inconsistent capture efficiency
  - Most SV breakpoints cannot be detected
  - More susceptible to GC-bias
  - Paired-end and split-read methods generally not well suited
  - Read depth methods
    - **Depth normalisation!!!**
  - E.g. CoNIFER,XHMM, ExomeCNV, VarScan2, SeqGene



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# Amplicon sequencing

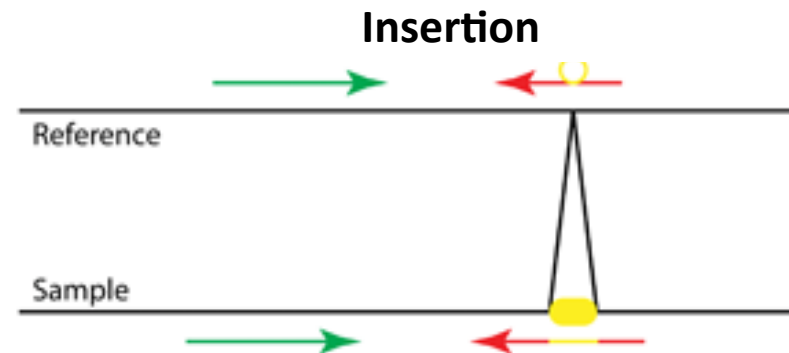
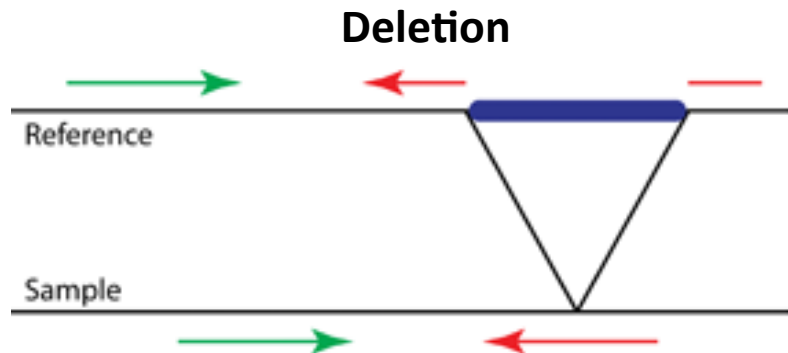
- Characteristics of the data
  - Different biases compared to WGS or WES data
  - Normalisation less effective due to limited target regions
  - High depth but very heterogeneous
  - GC-bias and amplicon length bias
  - Read depth methods
    - **Depth normalisation!!!**
  - E.g. ONCOCNV, AMS

## Practical:

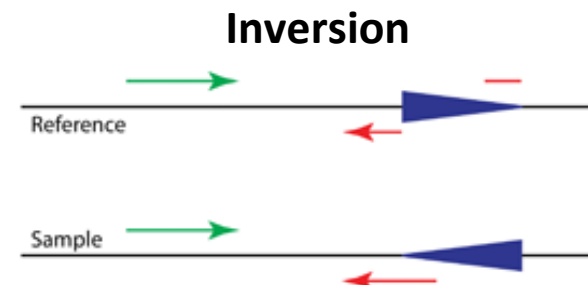
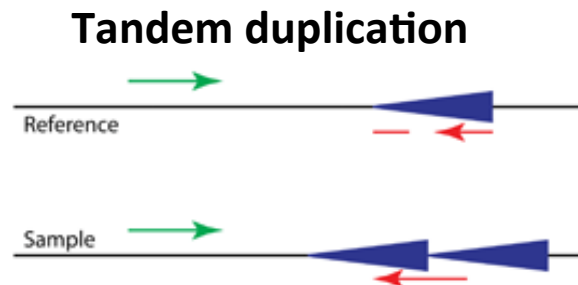
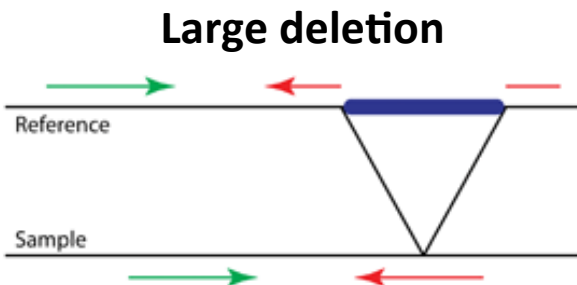
1. Obtain the insert size metrics for the samples using picard
2. Use Pindel to call a range of SVs
3. Conversion of pindel calls into vcf with some filtering using pindel2vcf

# Pindel - A pattern growth approach, split-read method

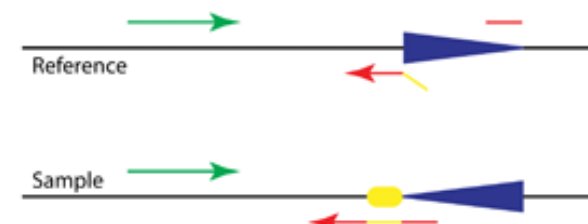
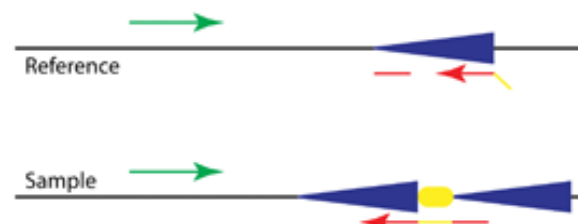
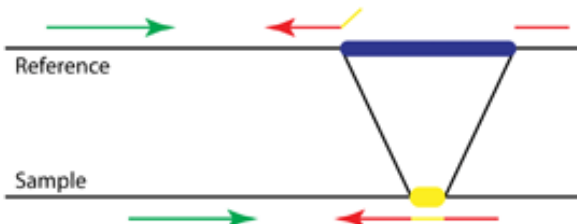
## Simple events:



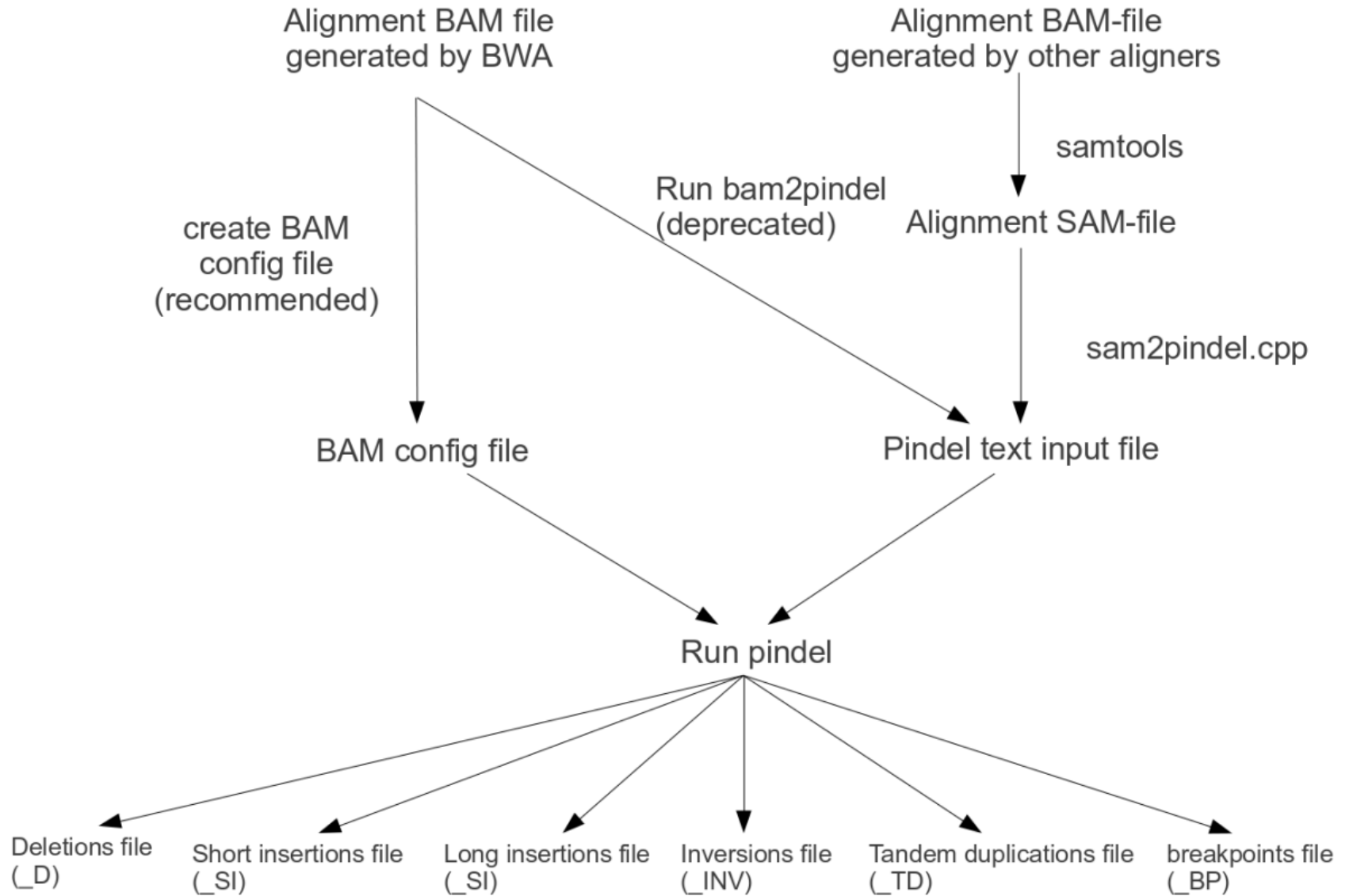
## Complex events:



## With additional sequence



## Pindel workflow:



## Make a new folder and copy the data:

```
cd /pico/scratch/usertrain/your_username/
mkdir SV
cd SV
```

```
cp /pico/scratch/userexternal/phallast/SV/bams/*.bam .
cp /pico/scratch/userexternal/phallast/SV/bams/*.bai .
cp /pico/scratch/userexternal/phallast/project2/ref/ucsc.hg19.fasta .
cp /pico/scratch/userexternal/phallast/project2/ref/ucsc.hg19.dict .
cp /pico/scratch/userexternal/phallast/project2/ref/ucsc.hg19.fasta.fai .
```

### The folder contains:

- bam files from two whole-genome sequenced chimpanzees Vaillant and Koby, sequenced on HiSeq 2000, with 100bp paired-end reads
- The bam files contain ~2.2Mb region of the Y chromosome (chrY: 13870437-16095787)
- Raw data has already been mapped to human genome reference hg19 using bwa mem v0.7.1 followed by standard bam refinement
- From Prado-Martinez et al "Great ape genetic diversity and population history." Nature. 2013 Jul 25;499(7459):471-5

## Obtain the insert size metrics for the samples using picard

```
module load profile/advanced
```

```
module load autoload picard/1.119
```

```
java -Xmx1G -jar /cineca/prod/applications/picard/1.119/binary/  
bin/CollectInsertSizeMetrics.jar I=PTV_Koby_Sk3_srt.bam  
O=PTV_Koby_Sk3_srt_InsMetr.txt H=PTV_Koby_Sk3_srt_hist.pdf
```

```
ls -l
```

```
more PTV_Koby_Sk3_srt_InsMetr.txt
```

Run the same for the other sample.

What is the mean insert size for these two samples?

Parameters used to run picard:

I Input bam file

O Output file containing Insert size metrics

H insert\_size\_Histogram.pdf

## Create a configuration file for pindel

Use a text editor to create a text file in the following format (separated by tabs):  
on each line, list the name of the bam-file, the insert size, and the label for the sample.  
For example:

```
path-to-data/sample1.bam    236  sample1  
path-to-data/sample2.bam    324  sample2
```

## Run Pindel:

```
module load profile/advanced
module load autoload pindel/1.0
pindel -h
```

```
pindel -f /pico/scratch/usertrain/your_username/SV/
ucsc.hg19.fasta -i conf_file.txt -c chrY -o /pico/scratch/
usertrain/your_username/SV/2chimps_pindel
```

### Parameters used to run Pindel:

- f The reference genome sequences in fasta format
- i The bam config file
- c Which chr/fragment
- o Output prefix

Pindel will output calls for the following SVs:

- D = deletion
- SI = short insertion
- INV = inversion
- TD = tandem duplication
- LI = large insertion
- BP = unassigned breakpoints



Look at the Pindel raw output:

```
cd /pico/scratch/usertrain/your_username/SV/  
ls  
more 2chimps_pindel_D
```

[illegible]

- |          |   |
|----------|---|
| 1        | Index   |
| 2        | The type of indel/SV  |
| 3        | The length of the SV  |
| 4        | “NT” indicate that the next number is the length of non-template sequences inserted   |
| 5        | the length(s) of the NT fragment(s)   |
| 6        | the sequence(s) of the NT fragment(s)   |
| 7-8      | the identifier of the chromosome the read was found on  |
| 9,10,11  | BP: the start and end positions of the SV   |
| 12,13,14 | BP_range if the exact position of the SV is unclear   |
| 15       | “Supports”  |
| 16       | The number of reads supporting the SV   |
| 17       | The number of unique reads supporting the SV (without duplicate reads)  |
| 18       | +: supports from reads whose anchors are upstream of the SV   |
| 19,20    | total and unique number of supporting reads whose anchors are upstream of the SV  |
| 21       | -: supports from reads whose anchors are downstream of the SV   |
| 22,23    | total and unique number of supporting reads whose anchors are downstream of the SV  |
| 24,25    | S1: a simple score, (“# +” + 1)* (“# -” + 1)  |
| 26,27    | SUM_MS: sum of mapping qualities of anchor reads  |
| 28       | the number of different samples scanned   |
| 29,30,31 | NumSupSamples?: the number of samples supporting the SV, as well as the number of samples having unique reads supporting the SV   |
| 32+      | Per sample: total nr. of supporting reads with anchors upstream, total nr. of unique supporting reads with anchors upstream, total nr. of supporting reads with anchors downstream, total nr. of unique supporting reads with anchors downstream. |

```

23      D 4      NT 0 "" ChrID chrY      BP 13885817      13885822      BP_range 13885817      13885828      Supports 3      3      + 2      2      -
1      1      S1 6      SUM_MS 108      2      NumSupSamples 1      1      PTT_Vaillant 6 6 2 2 1 1      PTV_Koby 0 0 0 0 0 0
ATTGATTTAATAAAGATATTTAAAAACAATGCTGCTGTTTATTTAATATCATAGCTACAGACCTATCACTGATTAAATAGATTTAAAACCAGTGCCAATGtttaTTTATTACTTCTTTTCTTTCTTTCTTTCTTTCTTTT
TTTTTTTTTTTTTGGAGATGGAGTCTTGC
TCAGGCTGGAGTGCAGTGGCACAAATC

      TGCTGCTGTTTATTTAATATCATAGCTACAGACCTATCACTGATTAAATAGATTTAAAACCAGTGTC AATG      TTTATTTATTCTTTTCTTTTCTTTTCTTT
-      13886248      21      PTT_Vaillant      @HWI-ST700660_83:1:2204:18492:133654#0/1
AAAAACAATGCTGCTGTTTATTTAATATCATAGCTACAGACCTATCACTGATTAAATAGATTTAAAGCCAGTGTC AATG      TTTATTTATTCTTTTCTTTTCTTT
+      13885381      60      PTT_Vaillant      @HWI-ST700660_83:2:2205:19803:196102#0/1
TTAAAAACAATGCTGCTGTTTATTTAATATCATAGCTACAGACCTATCACTGATTAAATAGATTTAAAACCAGTGCCAATG      TTTATTTATTCTTTTCTTT
+      13885346      27      PTT_Vaillant      @HWI-ST0757_59:1:1208:2217:14278#0/2

```

# Output file format:

- Reference sequence
- Sequence of the read
- Whether the anchor read is upstream(+) or downstream(-)
- The position of the mapped half of the paired-end read
- Mapping quality of the mapped read
- Sample name
- Read ID

Browsing thought the deletions output, which seems to be the most common lenght of these events? Which is the longest one you can find?

How many deletions and insertions in total have been called from these samples?  
(hint: try the “tail” command)

## Use pindel2vcf (conversion of Pindel output to VCF format and some filtering of SVs):

For full information about pindel2vcf options do (or check the handbook):

pindel2vcf

Then convert and filter our small insertion calls.

```
pindel2vcf -r /pico/scratch/usertrain/your_username/SV/  
ucsc.hg19.fasta -R hg19 -d Feb2009 -p /pico/scratch/usertrain/  
your_username/SV/2chimps_pindel_SI -v /pico/scratch/usertrain/  
your_username/SV/2chimps_pindel_SI.vcf -c chrY -is 5 -b -e 5 -sr  
13870437 -er 16095787
```

Parameters used to run pindel2vcf:

- r The reference genome sequences in fasta format
- R The name and version of the reference genome
- d The date of the version of the reference genome used
- p The name of the pindel output file containing the SVs
- v The name of the output vcf-file
- c The name of the chromosome
- is The minimum size of events to be reported
- b Only report events that are detected on both strands
- e The minimum number of supporting reads to report a SV
- sr The start of the region of which events are to be reported
- er The end of the region of which events are to be reported

**Choice of filters  
depend on what you  
are looking for!**

## pindel results in vcf format

```
more 2chimps_pindel_SI.vcf
```

```
##fileformat=VCFv4.0
##fileDate=Feb2009
##source=pindel
##reference=hg19
##INFO=<ID=END,Number=1,Type=Integer,Description="End position of the variant described in this record">
##INFO=<ID=HOMLEN,Number=1,Type=Integer,Description="Length of base pair identical micro-homology at event breakpoints">
##INFO=<ID=PF,Number=1,Type=Integer,Description="The number of samples carry the variant">
##INFO=<ID=HOMSEQ,Number=.,Type=String,Description="Sequence of base pair identical micro-homology at event breakpoints">
##INFO=<ID=SVLEN,Number=1,Type=Integer,Description="Difference in length between REF and ALT alleles">
##INFO=<ID=SVTYPE,Number=1,Type=String,Description="Type of structural variant">
##INFO=<ID=NTLEN,Number=.,Type=Integer,Description="Number of bases inserted in place of deleted code">
##FORMAT=<ID=PL,Number=3,Type=Integer,Description="Normalized, Phred-scaled likelihoods for genotypes as defined in the VCF specification">
##FORMAT=<ID=GT,Number=1,Type=String,Description="Genotype">
##FORMAT=<ID=RD,Number=1,Type=Integer,Description="Reference depth, how many reads support the reference">
##FORMAT=<ID=AD,Number=2,Type=Integer,Description="Allele depth, how many reads support this allele">
#CHROM POS ID REF ALT QUAL FILTER INFO FORMAT PTV_Vaillant PTV_Koby
chrY 13890888 . C CTTCTT . PASS END=13890888;HOMLEN=2;HOMSEQ=TT;SVLEN=5;SVTYPE=INS GT:AD 0/0:0,4 0/0:0,0
chrY 13890888 . C CTTCTTT . PASS END=13890888;HOMLEN=2;HOMSEQ=TT;SVLEN=6;SVTYPE=INS GT:AD 0/0:0,0 0/0:0,5
chrY 13895420 . A ATATTTTATTTTATT . PASS END=13895420;HOMLEN=14;HOMSEQ=TATTTTATTTTATT;SVLEN=15;SVTYPE=INS G
T:AD 0/0:5,3 0/0:2,3
chrY 13898547 . T TATAAATAAA . PASS END=13898547;HOMLEN=20;HOMSEQ=ATAAATAAATAAATAAATAA;SVLEN=10;SVTYPE=INS GT:A
D 0/0:0,3 0/0:1,1
```

How many small insertion calls were in the raw file and how many are in the filtered vcf?

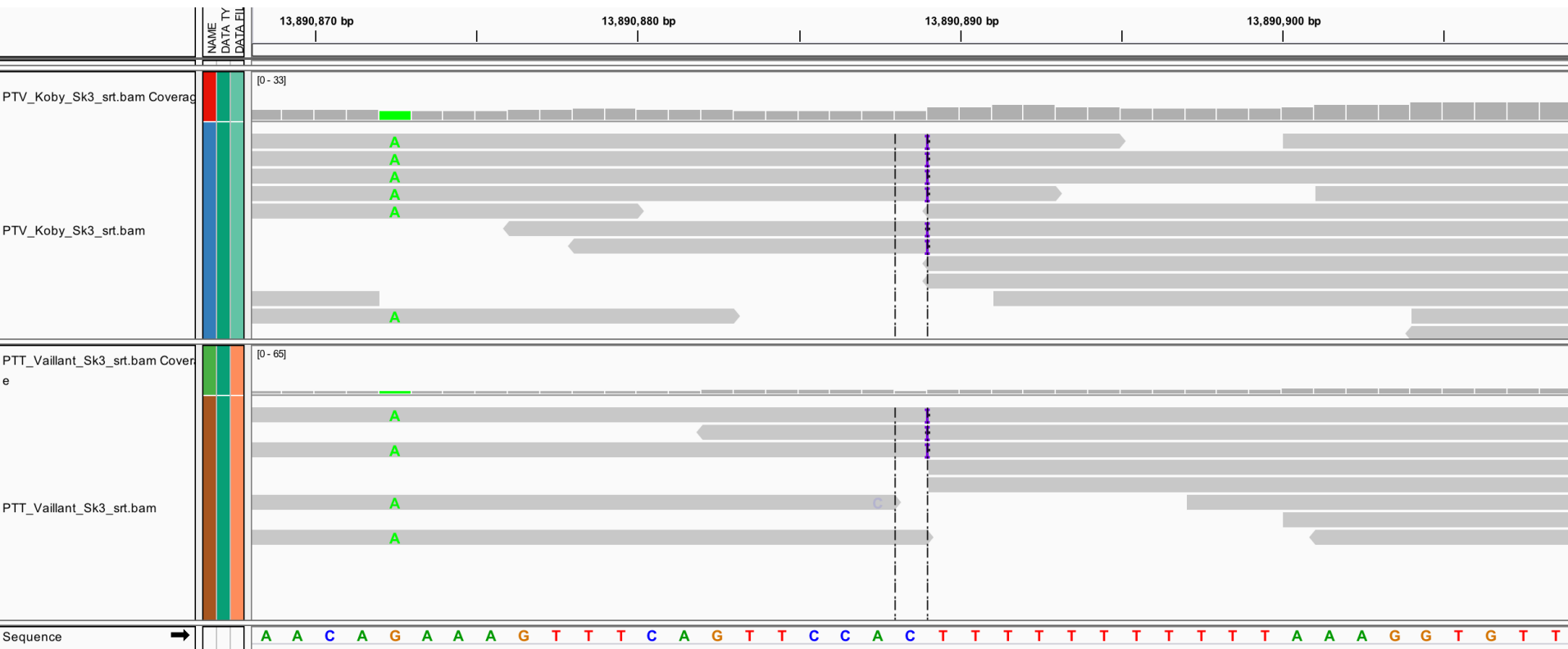
chrY 13890888 . C CTTCTT . PASS

PTT\_Vaillant PTV\_Koby

END=13890888;HOMLEN=2;HOMSEQ=TT;SVLEN=5;SVTYPE=INS GT:AD 0/0:0,4  
0/0:0,0

chrY 13890888 . C CTTCTTT . PASS

END=13890888;HOMLEN=2;HOMSEQ=TT;SVLEN=6;SVTYPE=INS GT:AD 0/0:0,0  
0/0:0,5



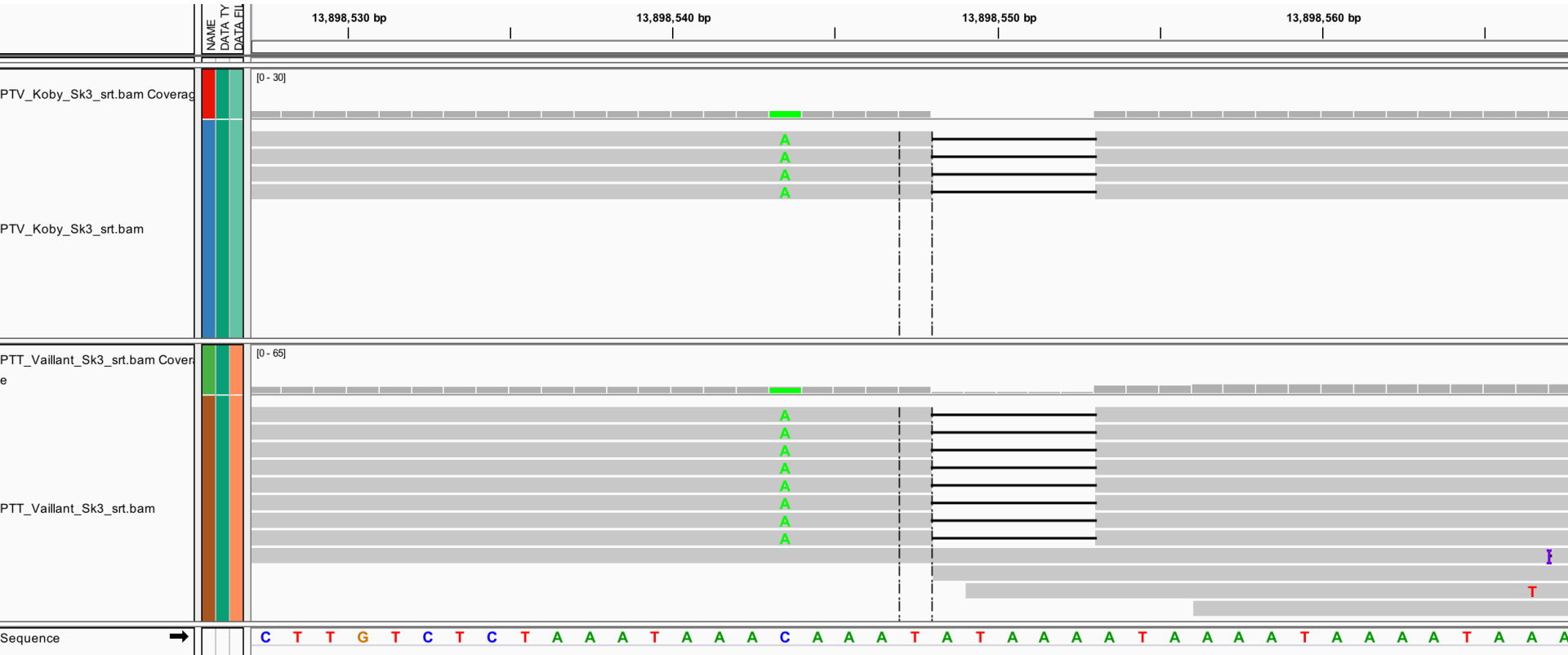
## Use pindel2vcf:

Convert and filter our deletion calls:

```
pindel2vcf -r /pico/scratch/usertrain/your_username/SV/  
ucsc.hg19.fasta -R hg19 -d Feb2009 -p /pico/scratch/usertrain/  
your_username/SV/2chimps_pindel_D -v /pico/scratch/usertrain/  
your_username/SV/2chimps_pindel_D.vcf -c chrY -is 3 -b -e 4 -sr  
13870437 -er 16095787
```

How many deletion calls were in the raw file and how many are in the filtered vcf?

chrY 13898547 . TATAAA T . PASS  
END=13898552;HOMLEN=15;HOMSEQ=ATAAAATAAAATAAAA;SVLEN=-5;SVTYPE=DEL  
GT:AD 1/1:1,10 0/0:1,4





chrY 15290968 . T n(6078) TGAGATGCAGTCTTGCTCTGTTGCCCAGATTGGAG .

PASS END=15297046;HOMLEN=0;SVLEN=-6078;SVTYPE=RPL;NTLEN=34

GT:AD 0/0:0,8 0/0:0,1



Choose a SV calling and filter them using different parameters.

Check the calls on IGV to get a feel of how the mapping of those positions looks.  
How reliable things look to you?